

The influence of environmental setting on the community ecology of Ediacaran organisms

Emily G. Mitchell¹, Nikolai Bobkov^{2,3}, Natalia Bykova^{2,4}, Alavya Dhungana,⁵ Anton Kolesnikov^{2,6, 7}, Ian R. P. Hogarth^{8,9} Alexander G. Liu¹⁰, Tom M.R. Mustill^{10,11}, Nikita Sozonov^{2,3}, Vladimir Rogov², Shuhai Xiao⁴ and Dmitriy V. Grazhdankin^{2,3}

ek338@cam.ac.uk

¹Department of Zoology, University of Cambridge, UK.

²Trofimuk Institute of Petroleum Geology and Geophysics, Novosibirsk, Russian Federation.

³Novosibirsk State University, Russian Federation.

⁴Department of Geosciences, Virginia Tech, Blacksburg, VA 24061, United States.

⁵Department of Earth Sciences, Durham University, UK.

⁶Geological Institute, Russian Academy of Sciences, Moscow, Russian Federation.

⁷Department of Geography, Moscow Pedagogical State University, Moscow, Russian Federation.

⁸Department of Chemical Engineering, University of Cambridge, UK.

⁹Current address: University College London, UCL Institute for Innovation and Public Purpose, UK.

¹⁰Department of Earth Sciences, University of Cambridge, UK.

¹¹Current address: Gripping Films, St. John's Church Road, London, England, E9 6EJ.

Abstract

The broad-scale environment plays a substantial role in shaping modern marine ecosystems, but the degree to which palaeocommunities were influenced by their environment is unclear. To investigate how broad-scale environment influenced the community ecology of early animal ecosystems, we employed spatial point process analyses (SPPA) to examine the community structure of seven bedding-plane assemblages of late Ediacaran age (558–550 Ma), drawn from a range of environmental settings and global localities. The studied palaeocommunities exhibit marked differences in the response of their component taxa to sub-metre-scale habitat heterogeneities on the seafloor. Shallow-marine (considered here as above storm wave-base) palaeocommunities were heavily influenced by local habitat heterogeneities, in contrast to their deeper-water counterparts. The local patchiness within shallow-water communities may have been further accentuated by the presence of grazers and detritivores, whose behaviours potentially initiated a chain of increasing habitat heterogeneity of benthic communities from shallow to deep-marine depositional environments. Higher species richness in shallow-water Ediacaran assemblages compared to deep-water counterparts across the studied time-interval could have been driven by this environmental patchiness, because habitat heterogeneities increase species richness in modern marine environments. Our results provide quantitative support for the “Savannah” hypothesis for early animal diversification – whereby Ediacaran diversification was driven by patchiness in the local benthic environment.

Keywords

Ediacaran, palaeoecology, spatial analysis, early animal diversification.

Author Contributions

E. Mitchell conceived this paper and wrote the first draft. N. Bobkov, A. Kolesnikov, N. Sozonov and D. Grazhdankin collected the data for DS surface. N. Bobkov and N. Sozonov

performed the analyses on DS surface. N. Bykova, S. Xiao, and D. Grazhdankin collected the data for KH1 and KH2 surfaces, and N. Bykova and D. Grazhdankin collected the data for WS-A, and E. Mitchell performed the analyses. A. Dhungana and A. Liu collected the data for FUN4 and FUN5 surfaces and A. Dhungana performed the analyses. T. Mustill and D. Grazhdankin collected the data for KS and T. Mustill and E. Mitchell performed the analyses. I. Hogarth developed the software for preliminary KS surface analyses. V. Rogov conducted the sedimentological study of the Khatyspyt Formation. E. Mitchell, N. Bobkov, N. Bykova, A. Dhungana, A. Kolesnikov, A. Liu, S. Xiao and D. Grazhdankin discussed the results and prepared the manuscript.

Background

The Ediacaran–Cambrian transition (~580–520 million years ago) is one of the most remarkable intervals in the history of life on Earth, witnessing the rise of macroscopic, complex animals in the global oceans (1,2). The diversification of early animals coincides with dramatic perturbations in the global environment, including changes to carbon cycling and a progressive but dynamic oxygenation of the oceans (3,4). The extent to which animals themselves drove these global changes is a matter of considerable debate (5–7). Competing hypotheses suggested to explain the observed environmental shifts include global abiotic changes that occurred over kilometre scales (8,9) and biotic factors acting over local scales (metre to kilometre), and include organism interactions such as burrowing and/or predation (10,11). Feedbacks between biotic and abiotic factors have also been proposed as drivers of early animal diversification, whereby Ediacaran organisms directly or indirectly created patchy food resources, stimulating the evolution of mobile bilaterians (12,13). Due to the small (within community) spatial scales over which key evolutionary mechanisms often act (14), investigation of the community ecology of Ediacaran assemblages from sites large distances (kilometres) apart offers an

75 opportunity to link the interactions of individual organisms to macro-evolutionary and macro-
76 ecological trends. In this study, we investigate the relationship between late Ediacaran early
77 animal diversification and the broad-scale environment.

78
79 Ediacaran macrofossils occur globally across a wide-range of palaeo-environments (1).
80 Previous studies have separated late Ediacaran palaeocommunities into three taxonomically
81 distinct assemblages – the Avalon, White Sea and Nama – which occupy partially overlapping
82 temporal intervals and water-depths, with no significant litho-taphonomic or biogeographic
83 influence (15–17). This study focusses on palaeocommunities within the Avalon and White
84 Sea fossil assemblages. Since these are considered to reflect original *in situ* communities
85 (18,19), they permit the use of statistical analysis of the distribution of fossil specimens on
86 bedding planes (spatial point process analyses, SPPA) to reconstruct the interaction of
87 organisms with each other and their local environment (20–25). The Avalon assemblage is
88 primarily represented by sites in Newfoundland (Canada) and Charnwood Forest (UK) (26,27),
89 and typically documents mid-shelf/deep-water settings (from depths below the storm-wave
90 base) of ~571–557 Ma (28,29). Such sites exhibit relatively limited ecological and
91 morphological diversity (30,31), and palaeocommunities consist almost exclusively of sessile
92 taxa (32) that show only weak trends in community composition along regional
93 palaeoenvironment gradients (20). Previous spatial analyses of Avalonian palaeocommunities
94 have found limited evidence for environmental interactions within these communities (21,23),
95 in contrast to the strong imprint exerted by resource-limitation on modern deep-sea ecosystems
96 (33,34).

97
98 Palaeocommunities from the White Sea assemblage are most famously represented by sites in
99 South Australia, and the East European Platform of Russia, dating to ~558–550 Ma (35–37).

100 These assemblages typically document shallow-water, diverse communities including some of
101 the oldest candidate bilaterians, with taxa interpreted to exhibit a wide range of ecological
102 strategies (7,12,36,38). Within the White Sea assemblages, community composition is
103 strongly correlated with facies and the presence of textured organic surfaces at bed-scale level
104 (39,40).

105
106 Metrics of taxonomic and ecological diversity are much higher in White Sea assemblages than
107 in Avalonian ones, with changes in taxonomic and morphological diversity calculated to be of
108 similar magnitude to those between the Ediacaran and Cambrian (30,31). These Ediacaran
109 assemblages have high beta-diversity compared to modern benthic systems (41), but the
110 driving processes underlying this high diversity are not understood. The regional
111 palaeoenvironment (kilometre scale) (15,17) has a significant influence on (non-algal
112 dominated) Ediacaran fossil assemblage composition, but its influence on local (metre to sub-
113 metre scale) community ecology has not yet been investigated. In modern benthic
114 communities, small spatial scale (< 50 cm) substrate heterogeneities (e.g. substrate variations
115 in nutrients, oxygen patchiness, or biotic and abiotic gradients within microbial mats) exert a
116 significant influence on community ecology (33,34,42). For Ediacaran palaeocommunities, it
117 is not possible from spatial analyses alone to determine the underlying causes of habitat
118 heterogeneities, nor the extent to which they relate to food resources, such as those resulting
119 from the decay of Ediacaran organisms (12,43). However, it is possible to compare how the
120 relative influence of such heterogeneities changes with broad-scale environmental setting:
121 previous analyses have identified assemblage-level trends between community compositions
122 and bathymetric depth (15–17). In this study, we compare the drivers of community ecology
123 between shallow and deep-water Ediacaran palaeocommunities (above or below the storm-
124 wave-base) over a ~8 million-year period using spatial analyses of seven palaeocommunities.

125

126 **Spatial analyses**

127 Determining the nature of interactions between fossilised organisms and their environment can
128 be undertaken if entire palaeocommunities are preserved *in situ*, such that the position of a
129 fossil on a bedding plane can be interpreted to reflect aspects of the original organism's life-
130 history (44). For sessile organisms, such as in the Avalon communities, community-scale
131 spatial distributions are dependent upon the interplay of a limited number of factors: physical
132 environment (which manifests as habitat associations of a taxon or taxon-pairs (45)); organism
133 dispersal/reproduction (46); competition for resources (47); facilitation between taxa (where
134 one taxon increases the survival another taxon) (48); and differential mortality (49). For fossil
135 assemblages containing mobile taxa (e.g. the White Sea assemblages), behavioural ecology
136 also influences spatial distributions, so interpretations of their spatial distributions are
137 necessarily qualitative rather than quantitative.

138

139 Studies of modern ecosystems have demonstrated that habitat associations resulting from
140 interactions between organisms and their local environment can be either positive, leading to
141 aggregations of individuals (such as around a preferential substrate for establishment), or
142 negative segregation away from such patches (21). SPPA are a suite of analyses which
143 compare the relative density of points (in this case fossil specimens) to different models
144 corresponding to different ecological processes, in order to infer the most likely underlying
145 process responsible for producing the observed spatial distribution (44,50). The application of
146 SPPA to Ediacaran palaeocommunities is documented in detail in the Methods section (see
147 also 21–23,25,50). For sessile organisms, habitat associations identified by SPPA are best-
148 modelled by a heterogeneous Poisson model (HP), or when combined with dispersal
149 limitations, an Inhomogeneous Thomas Cluster model (ITC) . Where the local environment is

resource-limited to the extent that it significantly reduces organism densities, this is indicated by spatial segregation between specimens within a community . When sessile populations are not significantly affected by their local environment, their spatial distributions are completely spatially random (CSR), indicating no significant influence by any biological or ecological processes at the spatial scale investigated (45,52–55). Alternatively, sessile populations which are not CSR and not affected by their local environment reflect dispersal/reproductive processes (45,52–55). CSR is modelled by homogeneous Poisson processes (44), whereas dispersal patterns are best modelled by best-fit Thomas Cluster (TC) or Double Thomas Cluster (DTC) models (52). Facilitation (where one taxon increases the survival of another) is best-modelled by linked-cluster models and density-dependent processes are detected using random-labeling analyses .

Geological setting

We assessed the community palaeoecology of seven fossil-bearing assemblages across five different global Ediacaran locations (Table 1, Figs 1, S1), spanning a range of habitats (Fig. S2) inhabited by members of the Ediacaran soft-bodied macrobiota during the late Ediacaran interval. Data from these assemblages were compared with seven palaeocommunities that have been subjected to SPPA in previous studies (see (21,23) for details of data collection and locality information). The diverse local depositional environments represented by these combined localities are here coarsely grouped within either shallow or deep-water settings, with deep-water defined as those surfaces below storm-wave base, to permit us to focus attention on the broadest macro-ecological and macro-evolutionary patterns in the data.

Shallow marine settings

Five of the studied palaeocommunities are found in facies that reflect a relatively shallow marine wave and current-agitated depositional environments. Palaeocommunity WS-A (White Sea) is an *Aspidella*-bearing surface (Fig. 1E) on the underside of a 15–30-mm-thick wave-rippled sandstone which forms part of a thick package of alternating wave-bedded sandstones, siltstones and mudstones interpreted as a product of progressive sediment-sorting by waves within a prograding, storm-influenced shoreface depositional system (Fig. S2; 55,56). It was collected from the Lyamtsa Formation of the Valdai Group, along the Onega Coast of the White Sea, Russian Federation. The original complete surface was studied in the field, where it has since been destroyed by landslides. *Aspidella* specimens were collected and are stored uncatalogued at the Trofimuk Institute for Petroleum Geology and Geophysics in Novosibirsk. The Lyamtsa Formation is older than a date of 558 ± 1 Ma (U/Pb zircon dating of volcanic tuffs near the base of the overlying Verkhovka Formation) (16).

Surface (KS) (Fig. 1A) originates from a section comprising 1.3–3.2 m thick, laterally continuous fining-upward sequences. Each sequence begins with channel casts or thick (0.5–0.6 m) packages of laterally discontinuous fossiliferous thin-bedded sandstones often exhibiting soft-sediment deformations (Fig. S2). These are followed by a package (0.4–0.7 m) of interbedded thinner wave-rippled sandstones, progressively thinning towards the top of the package. The upper part of each sequence is represented by an interval of alternating siltstone and shale. This section is interpreted as a prograding flood-influenced prodelta depositional system, and is part of the lower member of the Erga Formation (Winter Coast of the White Sea) (16,35), which is younger than 552.85 ± 0.77 Ma (58) (date recalculated from Martin et al. (59)). The KS surface was documented in the field and has been subsequently destroyed by landslides and weathering.

Two *Funisia*-bearing surfaces (Fig. 1D) from oscillation-rippled quartz-sandstones (the ‘ORS’ facies of (60)) are interpreted to have been deposited between fair-weather and storm-wave base under oscillatory and combined flow in the Ediacara Member of South Australia (39,61–63). These surfaces reside in the collections of the South Australia Museum, with surface FUN4 collected from Ediacara Conservation Park (SAM P55236) and surface FUN5 collected from the Mount Scott Range (SAM P41506). Since FUN4 and FUN5 originate from different localities (> 50 km apart), it is assumed likely that they represent discrete bedding plane palaeocommunities. The South Australian Ediacaran successions have not been radiometrically dated, but the Ediacara Member is widely assumed to be of a similar age to the Russian White Sea fossil-bearing sections (1,2).

Surface DS is a *Dickinsonia*-bearing surface (Fig. 1B) from the Konovalovka Member of the Cherny Kamen Formation, cropping out along the Sylvitsa River, Central Urals, Russian Federation (64,65). It lies within an interval of finely alternating wave-rippled sandstones, siltstones and mudstones that are sandwiched between two thick intervals of biolaminated sandstone characterised by microbial shrinkage cracks and salt crystal pseudomorphs (66). The overall succession is considered transitional from marginal marine to non-marine, with the fossil-bearing interval interpreted as having been deposited in a lagoon within a tidal flat depositional system (66). A U/Pb zircon date of 557 ± 13 Ma from volcanic tuffs near the base of the Cherny Kamen Formation (65) suggests that this unit may have been deposited broadly coevally with those on the White Sea coast. Specimens from this surface reside in Novosibirsk State University, Russian Federation (specimen numbers: 2057-001 to 2057-003) and will be placed at the Ural Geological Museum (Yekaterinburg).

All five of these surfaces, therefore, represent siliciclastic depositional environments from above the storm-wave base (Fig S2), and so fall broadly into the grouping of “shallow

marine”. They contain examples of taxa interpreted as animals (e.g. *Dickinsonia* (67), *Kimberella* (68)) as well as non-metazoans (*Orbisiana*) (69), and their age and facies place them within the White Sea assemblage (15,17).

Deep marine settings

Two bedding surfaces dominated by *Aspidella* specimens (KH1 and KH2, Fig. 1C) were collected from a package of finely alternating limestone and shale interbeds within the uppermost Khatyspyt Formation, Khorbusuonka River, Siberia. The Khatyspyt Formation records a relatively narrow basin with steep slopes developed in a marine ramp setting (Fig. S2). Sedimentological observations (e.g., turbiditic nature of the limestones; evidence of strong unidirectional flows; intraclasts originating from outside of the Khatyspyt depositional basin) suggest the Khatyspyt Formation was deposited within a starved intracratonic rift, beyond the flexure slope break (70–73). The lack of any evidence of wave reworking within the entire Khatyspyt Formation, and the geochemical evidence of stratification and anoxia (74), with episodic euxinia, corroborate a deep-water interpretation of the Khatyspyt depositional environment. A positive $\delta^{13}\text{C}_{\text{carb}}$ excursion in the Khatyspyt Formation has been correlated with an excursion of similar magnitude in the <550 Ma Gaojiashan Member of the Dengying Formation (73). Strontium isotope ratios ($^{87}\text{Sr}/^{86}\text{Sr}$) in the Khatyspyt Formation are consistently ca. 0.7080 (73,74), a value approaching some of the ratios seen in the Gaojiashan Member (75), so this correlation seems plausible. Surface KH2 remains in the field, and surface KH1 was destroyed by excavating surface KH2. Specimens from surface KH1 reside in Trofimuk Institute for Petroleum Geology and Geophysics, collection number 913 (specimen numbers: 0607/2009-3, 0607/2009-6, 0607/2009-7, 0607/2009-17, 0607/2009-18).

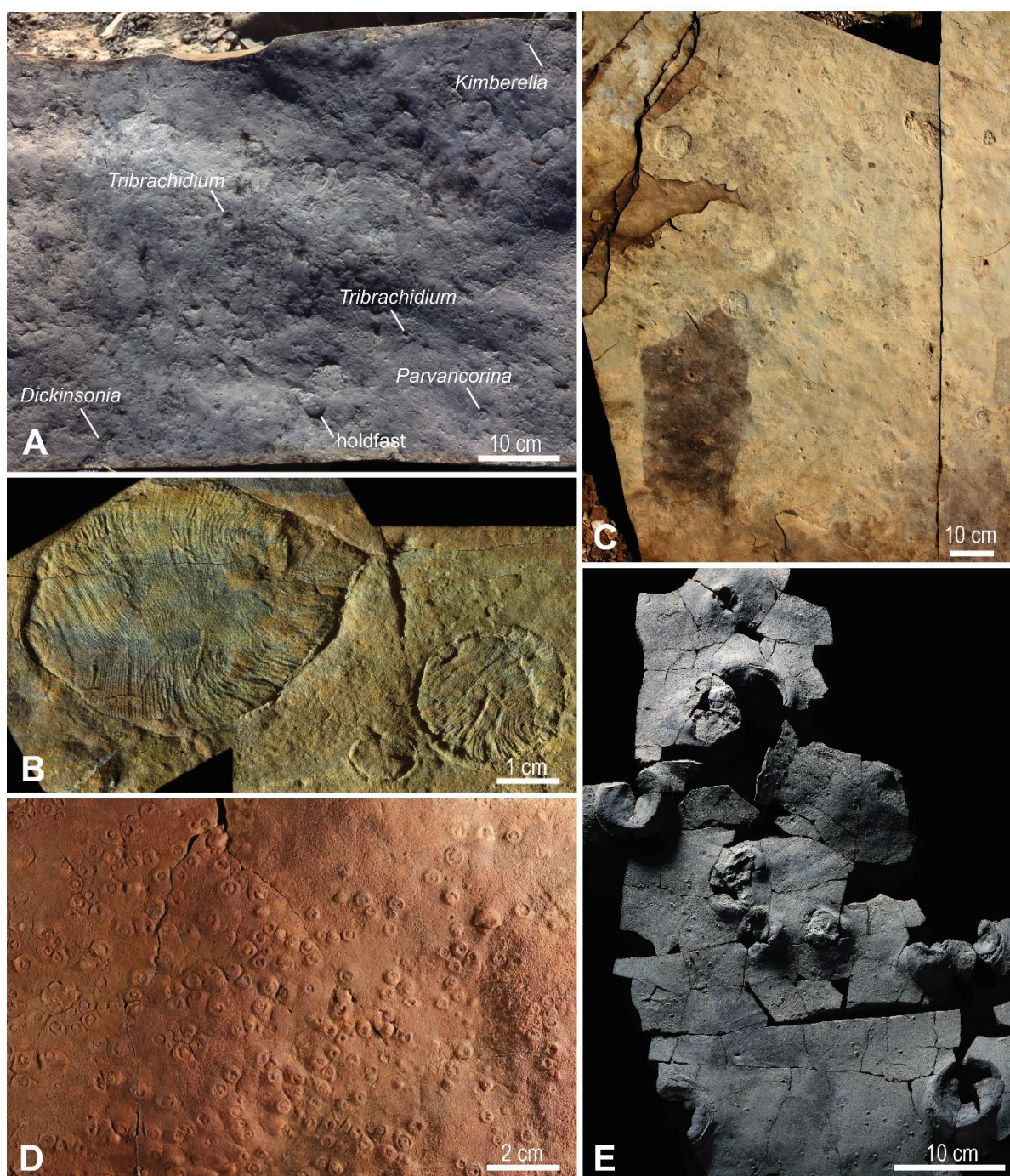


Fig. 1. Assemblages of Ediacaran fossils from study localities. A) A fragment of the *Kimberella* surface (KS), indicating key taxa, lower Erga Formation, Winter Coast of the White Sea. B) Specimens of *Dickinsonia* from the *Dickinsonia* surface (DS), Konovalovka Member, Cherny Kamen Formation, Sylvitsa River, Central Urals. C) A representative fragment of the *Aspidella* surface (KH1), Khatyspyt Formation, Olenek Uplift, Northern Siberia. D) *Funisia* from FUN4 surface (SAM P55236), Ediacara Member, Rawnsley Quartzite, South Ediacara

Conservation Park, Flinders Ranges, South Australia. E) A representative fragment of the WS-A surface, upper Lyamtsa Formation, White Sea Region. This particular fragment was not included in the analysis.

Data Collection

Spatial data were collected from the surfaces using different methods depending on the physical properties of the bedding plane (Table 1). The WS-A, KH1, KH2 surfaces were mapped in the field (WS-A in 2017, KH1 in 2006 and 2009, and KH2 in 2018) onto millimetre graph paper (Fig. 2). First, the co-ordinates of the edge of the rock surface were recorded, then the co-ordinates, orientation and dimensions of each of the specimen were measured and plotted onto the paper. For DS, a bedding surface of 9 m² was excavated over the course of two years (2017–2018). The surface was photo-mapped, with photographs taken under an artificial light source at night. The intersection between maximum length (L) and maximum width (W) of each specimen was taken to be the absolute position of the organism, with measurements obtained from digital photographs using Adobe Photoshop CC software and Apple Script Editor (Fig. 2).

Surface	Environmental setting	Species richness	Dominant Taxon/taxa	Specimen numbers	Area mapped (m ²)
WS-A	Shallow	1	<i>Aspidella</i>	40	0.54
KH1	Deep	2	<i>Aspidella</i>	204	2.38
KH2	Deep	2	<i>Aspidella</i>	81	1.52
DS	Shallow	1	<i>Dickinsonia</i>	62	9.00
KS	Shallow	13	<i>Kimberella</i> , <i>Orbisiana</i>	107	2.74
FUN4	Shallow	2	<i>Funisia</i>	290	0.69
FUN5	Shallow	1	<i>Funisia</i>	482	0.78

Table 1. Summary data of the surfaces mapped. The environmental setting, species richness, specimen numbers within the mapped area, and the total mapped area are provided.

The KS surface was excavated in July 2004, and is a laterally discontinuous transect consisting of four slabs of variable size, ranging from 0.6×0.4 m to 1.6×1.0 m. The relative positions of the slabs within the transect were mapped *in situ* on an excavated terrace. A separate block originating from the same horizon was found in float close to the transect. Following reassembly, the taxonomic identity, positions, orientations and shapes of the fossils were mapped at millimetre scale (Fig. 2). For the FUN4 and FUN5 surfaces, photogrammetric maps of the bedding surfaces were made, with lens edge effects corrected using RawTherapee (v. 2.4.1; Fig. 2). For all mapped palaeocommunities, fossil identification, position, and dimensions (disc width, disc length, stem length, stem width, frond length, and frond width) were digitized in Inkscape 0.92.3 on a 2D projection of the dataset, resulting in a 2D vector map for each palaeocommunity (Fig. 2). Only taxa that had sufficient abundance (> 5 specimens) for spatial analyses were formally identified, and these were grouped within one of six taxonomic groups: *Aspidella*, *Dickinsonia*, *Funisia*, *Kimberella*, *Orbisiana*, and the trace fossil *Kimberichnus*. *Aspidella* is considered a form taxon here, in line with previous studies (66,80), and may represent multiple different types of organism, such as the holdfasts of frondose taxa. We also note that on FUN4 and FUN5, *Funisia* fossils are represented only by their holdfast ‘buds’, rather than by the complete *Funisia* tubular organism (76). A group consisting of all the sessile taxa on the KS surface was also assessed since abundance was not sufficient to include all taxa individually. Taxa were excluded from analyses if there were fewer than five specimens on the surface, because low abundance taxa would fall below the threshold for which results would be statistically meaningful.

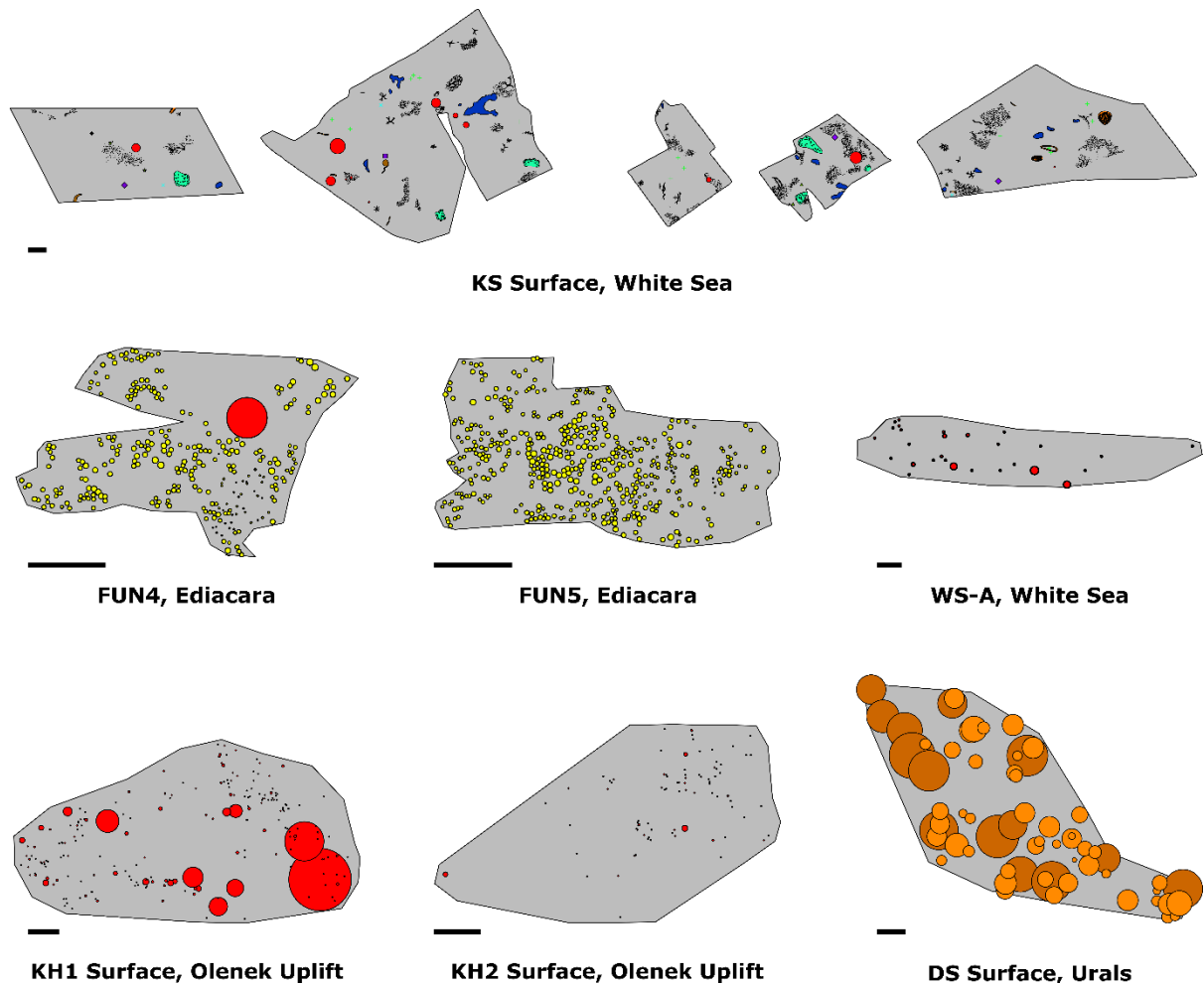


Fig 2. Spatial maps of the seven studied palaeocommunities. Scale bar = 10 cm. Different colours indicate different trace and body fossils as follows: Red, *Aspidella*; Orange, *Dickinsonia*; Yellow circles, *Funisia*; Light green scratch marks, *Kimberichnus*; Light green crosses, *Kimberella*; Blue crosses, *Charniodiscus*; Green triangles, *Parvancorina*; Dark blue patches, *Orbisiana*; Black stipples, horizontal traces; White globular strings, *Palaeopasichnus*; Purple diamonds, *Andiva*; Purple squares, *Yorgia*. Size of the circles corresponds to specimen length or diameter (as appropriate). On the DS surface, dark orange circles are the large size-class of *Dickinsonia*, and the light orange represents the small size-class. See Fig S3 for a high resolution image of the KS surface.

Methods

311 **Bias analyses**

312 For each surface, we first investigated erosional biases by testing to see if fossil density is
313 correlated with sources of modern erosion, such as distance of the surface from the ground, or
314 from a water source (18,72). Since the preserved palaeo-communities are a sub-sample of the
315 communities alive at the time, we would not expect specimen densities to correlate with
316 modern bedding-plane features unless these features were affecting the fossil distributions.
317 Tectonic deformation was tested for by inspection of specimen and bedding-plane deformation
318 (18,72). If these factors were found to have significantly affected specimen density
319 distributions, the erosion and/or deformation were taken into account when performing later
320 analyses cf (. 23), with heavily eroded sections of the bedding planes excluded from analyses.
321 The influence of tectonic deformation was only observed on the DS surface, so
322 retrodeformation techniques (18,25) were not applied to the spatial maps of WS-A, KH1, KH2,
323 KS, FUN4 and FUN5 surfaces. Where possible (WS-A, KH1 and KH2 surfaces), the area near
324 the outcrops was investigated, and no independent evidence of tectonic deformation was found.
325 The holdfast discs on surfaces KS, FUN4 and FUN5 did not show any evidence of tectonic
326 deformation. The DS surface showed signs of deformation in the form of consistent variation
327 in specimen length to width ratios along a presumed axis of deformation. The *fitModel*
328 function from the *mosaic* package in R (73) was used to find the best-fit values for the
329 direction and strength of deformation using the assumption that *Dickinsonia* had a consistent
330 length to width ratio during its ontogeny (40,77,78) though note (79), and the spatial map was
331 retrodeformed (cf. 18,23,25).

332

333 **Spatial Analyses**

334 Initial data exploration, inhomogeneous Poisson modelling, and segregation tests were

performed in R (80) using the package *spatstat* (81,82). Programita software was used to obtain distance measurements and to perform aggregation model fitting (described in detail in references (45,49,81,83–87).

Univariate and bivariate pair correlation functions (PCFs) were calculated from assemblage population densities using a grid of $1\text{ cm} \times 1\text{ cm}$ cells on all surfaces except DS, where a $10\text{ cm} \times 10\text{ cm}$ cell size was used to correspond to the larger overall mapped area. To minimise noise, a 3 cell smoothing was calculated dependent on specimen abundance, which was applied to the PCF. To test whether the PCF exhibited complete spatial randomness (CSR), 999 simulations were run for each univariate and bivariate distribution, with the 49 highest and 49 lowest values removed. CSR was modelled by a Poisson model on a homogeneous background where the $\text{PCF} = 1$ and the fit of the fossil data to CSR was assessed using Diggle's goodness-of-fit test. Note that due to non-independence of spatial data, Monte-Carlo generated simulation envelopes cannot be interpreted as confidence intervals. If the observed data fell below the Monte-Carlo simulations, the bivariate distribution was interpreted to be segregated; above the Monte-Carlo simulations, the bivariate distribution was interpreted to be aggregated.

If a taxon was not randomly distributed on a homogeneous background, and was aggregated, the random model on a heterogeneous background was tested by creating a heterogeneous background from the density map of the taxon under consideration. This density map was defined by a circle of radius R over which the density was averaged throughout the sample area. Density maps were formed using estimators within the range of $0.1\text{ m} < R < 1\text{ m}$, with R corresponding to the best-fit model used. If excursions outside the simulation envelopes

for both homogeneous and heterogeneous Poisson models remained, then Thomas cluster models were fitted to the data as follows:

1. The PCF and L-function (88) of the observed data were found. Both measures were calculated to ensure that the best-fit model is not optimized towards only one distance measure, and thus encapsulates all spatial characteristics.
2. Best-fit Thomas cluster processes (89) were fitted to the two functions where $PCF > 1$. The best-fit lines were not fitted to fluctuations around the random line of $PCF = 1$ in order to aid good fit about the actual aggregations, and to limit fitting of the model about random fluctuations. Programita used the minimal contrast method to find the best-fit model.
3. If the model did not describe the observed data well, the lines were re-fitted using just the PCF. If that fit was also poor, then only the L-function was used.
4. Ninety nine simulations of this model were generated to create simulation envelopes, and the fit checked using the O-ring statistic (83).
5. In order to assess how well the model fit the observed data, the goodness-of-fit (p_d) was calculated over the model range (87). A $p_d = 0$ indicates no model fit, and $p_d = 1$ indicates a perfect model fit. Very small-scale segregations (of the order of specimen diameter) were not included in the model fitting, since they likely represent the finite size of the specimens, and a lack of specimen overlap.
6. If there were no excursions outside the simulation envelope and the p_d -value was high, then a univariate homogeneous Thomas cluster model was interpreted as the best model.

For any univariate distributions exhibiting CSR, the size-classes of each taxon were calculated, the univariate PCFs of the smallest size-classes and largest size-classes were plotted, with 999 Monte Carlo simulations of a complete spatially random distribution and

segregation tests performed. The most objective way to resolve the number and range of size classes in a population is by fitting height-frequency distribution data to various models, followed by comparison of (logarithmically scaled) Bayesian information criterion (BIC) values (87), which we performed in R using the package MCLUST (90). The number of populations identified was then used to define the most appropriate size classes. A BIC value difference of >10 corresponds to a “decisive” rejection of the hypothesis that two models are the same, whereas values <6 indicate only weakly rejected similarity of the models (90–94). Once defined, the PCFs for each size class were calculated.

Bivariate analyses were performed on the KS surface (the only surface with multiple abundant taxa/taxon groups) between *Kimberella* – *Orbisiana*, *Kimberella* – *Kimberichnus* and *Orbisiana* – *Kimberichnus*. For each ‘taxon’ pair, the bivariate PCF was calculated, and then compared to CSR using Monte Carlo simulations and Diggle’s goodness-of-fit test.

Results

Across the seven palaeocommunities, *Dickinsonia* on the DS surface was the only taxon that exhibited CSR (Table 2). There were five univariate distributions (*Sessile Taxa* on KS, *Funisia* on FUN4 and FUN5, *Aspidella* on KH1 and KH2) exhibiting aggregated spatial distributions, and two univariate (*Aspidella* on WS-A and large *Dickinsonia* on DS) and one bivariate (*Kimberella* and *Kimberichnus* on KS) segregated spatial distributions (Fig. 3, Table 2). The *Aspidella* aggregations from KH1 and KH2 were best modelled by the same double Thomas cluster process ($p_d^{kh1} = 0.883$, $p_d^{k21} = 0.932$, Fig. 3G, H; Table 2), which consisted of large clusters of 20.96 cm diameter containing smaller clusters with a mean of six specimens within a cluster of 7.34 cm in diameter (Fig. 3G and H, (95)). These results indicate that the non-random spatial distributions were most likely due to two generations of reproduction cf. (44), and do not represent a significant interaction or association with local habitat variations. This

result is consistent with previous work on older (~565 Ma) deep-water communities that also show a strong non-environmentally influenced signal (23). In contrast, the *Aspidella* from the WS-A surface show significant segregation and are best-modelled by a heterogeneous Poisson process ($p_d^{WS-A} = 0.796$, Fig 3F, Table 2). This is consistent with small-scale intra-specific competition in a resource-limited environment (96). *Funisia* from FUN4 and FUN5 had aggregations that are best-modelled by heterogeneous Poisson processes ($p_d^{Fun4} = 0.9570$, $p_d^{Fun5} = 0.9080$, Fig. 3 D, E; Table 2), which are interpreted to indicate significant habitat associations with the local environment.

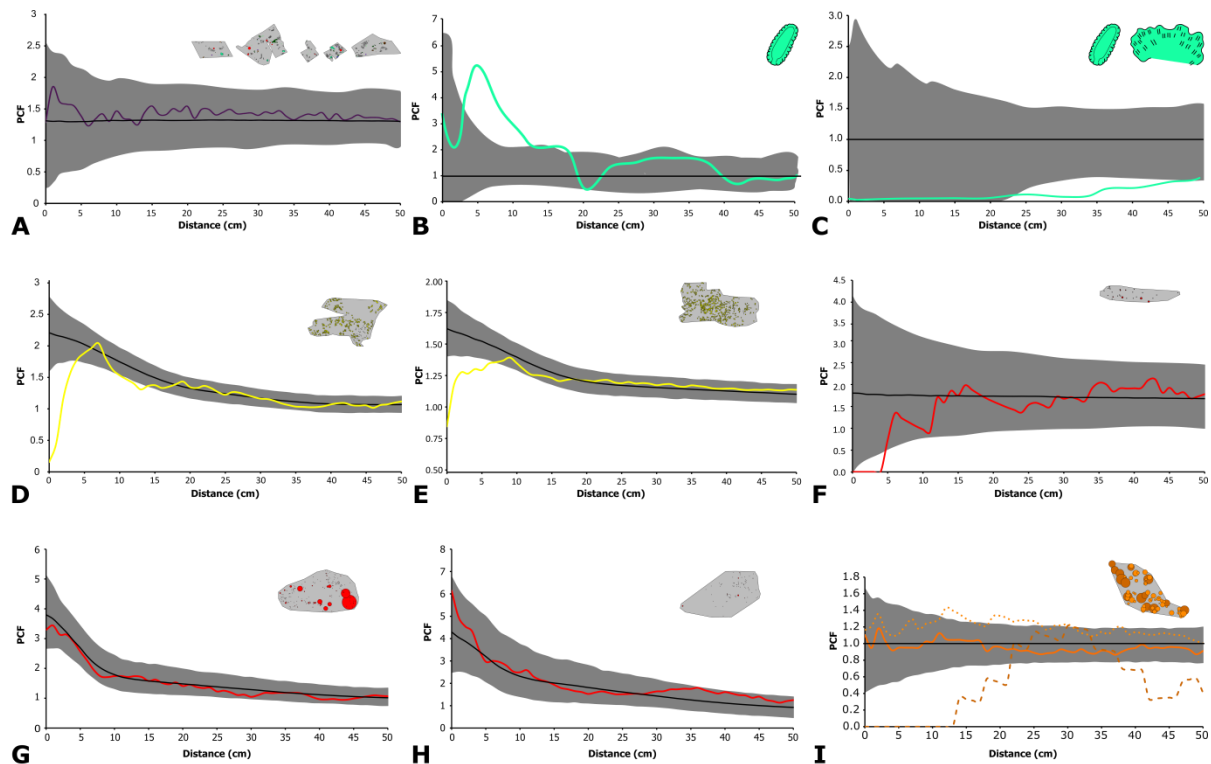


Fig. 3. Pair correlation functions describing the spatial distributions of the seven studied palaeocommunities. The coloured lines are the observed data and black lines represent best-fit models (either CSR or heterogeneous Poisson). The grey area is the simulation envelope for 999 Monte Carlo simulations. The x-axis is the inter-point distance between organisms in centimetres. On the y-axis, PCF = 1 indicates complete spatial randomness (CSR), < 1 indicates segregation, and > 1 indicates aggregation. A) The KS surfaces showing sessile specimens with

the black-line showing the best-fit heterogeneous Poisson model. B) KS univariate *Kimberella*. C) KS bivariate *Kimberella* – *Kimberichnus* with the CSR model shown. D) FUN4, and E) FUN5 surfaces showing the *Funisia* distributions with the best-fit heterogeneous Poisson model. *Aspidella* from F) WS-A, G) KH1 and H) KH2 surfaces with their best-fit heterogeneous Poisson models. I) *Dickinsonia* from DS with the solid line showing the whole population, dotted line the juveniles and dashed line the adults with the CSR model shown.

SURFACE	TAXON	N	<i>P</i> values				
			CSR	HP	TC	DTC	ITC
WS-A	<i>Aspidella</i>	40	0.019	0.796	0.504	0.2759	0.425
KH1	<i>Aspidella</i>	204	0.001	0.001	0.648	0.883	0.313
KH2	<i>Aspidella</i>	81	0.001	0.001	0.576	0.932	0.001
FUN4	<i>Funisia</i>	290	0.001	0.9570	0.6340	NA	0.245
FUN5	<i>Funisia</i>	482	0.001	0.9080	0.1320	NA	0.218
DS	<i>Dickinsonia</i>	62	0.857	0.022	0.025	NA	0.019
	<i>Dickinsonia</i> Small	48	0.128	0.978	0.143	NA	0.158
	<i>Dickinsonia</i> Large	14	0.388	0.446	0.409	NA	0.434
KS	All	107	0.858	0.381	0.328	NA	0.380
	All sessile	44	0.033	0.956	0.770	NA	0.761
	<i>Kimberella</i>	18	0.001	0.837	0.491	NA	0.103
	<i>Orbisiana</i>	16	0.325	0.332	0.326	NA	0.288
	<i>Kimberichnus</i>	6	0.566	NA	NA	NA	NA
	Bivariate <i>Kimberella</i> – <i>Kimberichnus</i>	24	0.028	NA	NA	NA	NA

Table 2. Goodness-of-fit tests for spatial analyses. For the inhomogeneous point processes (HP and ITC), the moving window radius is 0.5 m, using the same taxon density as the taxon being modelled. $p_d = 1$ corresponds to a perfect fit of the model to the data, while $p_d = 0$ corresponds to no fit. Where observed data did not fall outside CSR Monte-Carlo simulation envelopes, no further analyses were performed, which is indicated by NA. CSR: Complete spatial randomness; HP: Heterogeneous Poisson model; TC: Thomas cluster model; DTC: double Thomas Cluster; ITC: Inhomogeneous Thomas cluster model. N is the number of specimens mapped. Note that for the mobile taxa *Dickinsonia* and *Kimberella*, and presumed trace fossils formed by mobile taxa (*Kimberichnus*), the observed spatial pattern will also be defined by their behaviour, and so the inference of process from pattern is not as

straightforward (see discussion in the main text). The p_d -value of the best-fit model is given in bold.

The KS community is notably different in species composition from deep-water communities because it contains mobile organisms such as *Kimberella* and *Yorgia* (97–100) as well as putative trace fossils such as *Kimberichnus* (thought to be produced by the grazing activity of *Kimberella* specimens) (101). We found that the KS community exhibits CSR, which suggests that any taxon-specific univariate distributions are likely to be biological/ecological in origin, rather than resulting from a taphonomic bias ($p_d^{KS_{All}} = 0.858$, Table 2, (23)). In contrast, when all the sessile taxa were grouped together they exhibited a significant aggregation (Table 2), which was best-modelled by a heterogeneous Poisson process ($p_d^{KS_{Sessile}} = 0.956$, Table 2). *Kimberella* exhibits a significant aggregation under spatial scales of 20 cm ($p_d^{KS_{Kimberella}} = 0.001$ for CSR model, Fig. 3A), with Thomas cluster and heterogeneous Poisson models fitting the data well, suggesting that behavioural factors may also influence *Kimberella* spatial patterns. The *Kimberichnus* PCF spatial distribution has a CSR distribution (Fig. 3B, $p_d^{KS_{Rad}} = 0.566$, Table 2). Furthermore, the bivariate analyses between *Kimberella* and *Kimberichnus* show a significant segregation ($p_d^{KS_{KimRad}} = 0.028$, Fig. 3C), which could reflect the *Kimberella* organisms avoiding patches of the surface that had already been grazed.

The *Dickinsonia* population from DS exhibited a CSR PCF distribution (Fig. 3I, $p_d = 0.857$). Previous analysis of the population of *Dickinsonia* from DS showed two cohorts in the size-distribution (95). The two cohorts exhibited different PCF spatial behavior, with the small specimens aggregating with a best-fit heterogeneous Poisson model (Fig. 3I, $p_d^{small} = 0.978$) and the large specimens exhibiting segregation (Fig. 3I).

Interpreting the spatial distributions of mobile organisms

For mobile organisms, inferring the underlying process behind the observed spatial distributions is imprecise, since their spatial patterns also incorporate contributions from their behaviour. Modern animals move primarily to find resources, mates, microhabitats and/or escape predators or detrimental environmental conditions. There is no direct evidence of predators in the Ediacaran, unless perforations in the terminal Ediacaran calcareous tubes of *Cloudina* could be referred to as borings (102). The lack of resolution of reproductive strategies of *Dickinsonia* means that we cannot predict resultant spatial patterns, and thus cannot definitively rule reproductive processes out as a source of spatial patterning. Such reproduction may or may not include necessity for close proximity and/or broadcast spawning cf. the majority of benthic organisms (103). However, reproductive processes are considered unlikely explanation for the observed spatial patterns because the *Dickinsonia* do not aggregate as might be expected in a mating event as demonstrated by the univariate of the largest size-class in the studied *Dickinsonia* population segregation (Fig. 3C, Table 2). Furthermore, broadcast spawning does not require the two mating organisms to be within the spatial scale (< 40 cm) found on the DS surface. We cannot determine whether the large *Dickinsonia* are reacting to the mortality event that killed and preserved them, however, this would not explain the complex interplay between aggregation and segregated behaviors. Therefore, for this *Dickinsonia* population, the search for resources and/or microhabitats is considered the most plausible explanation, particularly since this hypothesis is further supported by their spatial patterns. Aggregated – segregated PCF patterns such as those seen in our *Dickinsonia* population are common in extant sessile organisms where juveniles are initially aggregated on preferred habitats but then begin to compete with each other as they require greater resources, leading to thinning or segregation amongst adult populations. While it is not possible to confirm the underlying mechanism for the distribution of the studied *Dickinsonia* population,

given the preceding points we consider it most likely to be motivated by associations with preferential habitat for food and/or resources. Further analyses of other *Dickinsonia* surfaces would enable more robust conclusions to be reached.

Time averaging

The preservation of time-averaged communities has the potential to bias our analyses (see (21,25). In Avalonian communities, taphomorphs interpreted to record the decaying remains of organisms are identified by their poor preservational fidelity, irregular morphologies, and often high topographic relief (104). This interpretation is consistent with data suggesting that the spatial interactions of some taphomorph populations mirror those of other taxa they are considered to be derived from (21). Taphomorphs are considered unlikely to have imparted a significant signal on these studied surfaces, since we did not observe ivesheadiomorph-type forms, and there is a consistent level of preservational detail amongst fossil communities

Funisia communities tend to have very similar diameters for the holdfasts, which suggests single colonization events (76). Different reproductive events can be distinguished by population analyses of size-distributions (105), with each reproductive event identified through statistically significant cohorts within the size-distribution (90). Surfaces FUN4 and FUN5 both exhibit populations with two cohorts (Fig. S4), most likely indicating two reproductive/colonization events. The best-fit models for each of these surfaces are heterogeneous Poisson models (Fig. 3, Table 2), with very high goodness-of-fit values ($p_d > 0.90$) reflecting a single model for each surface. Therefore, cohorts of *Funisia* specimens on each of the studied surfaces were affected by the same underlying environmental heterogeneity, so most likely were contemporaneous.

Discussion

The univariate and bivariate analyses exhibited by taxa within of five out of seven of the studied palaeocommunities provide compelling evidence that their local environment had a significant influence on their communities (Fig. 3, Table 2). In modern settings, habitat associations form when a patchy resource provides heterogeneously distributed preferential conditions for the establishment and growth of sessile taxa, and/or feeding ‘hotspots’ for mobile taxa . The presence of inferred habitat interactions within our palaeocommunities showed a significant correlation with the environmental setting (Kruskal-Wallis Test, $p = 0.049$), with all five palaeocommunities with strong habitat interactions derived from depositional environments that we consider to be shallow-water settings. The two communities that were seemingly not strongly influenced by their local habitat are from deep-water facies (Table 2). These results are consistent with previous work, which found that for seven independent deep-marine (slope and basin) Ediacaran palaeocommunities from Newfoundland and Charnwood Forest, only one was dominated by associations of taxa with local habitat heterogeneities (21–23) (Kruskal-Wallis Test of all data, $p = 0.021$; Fig. 4).

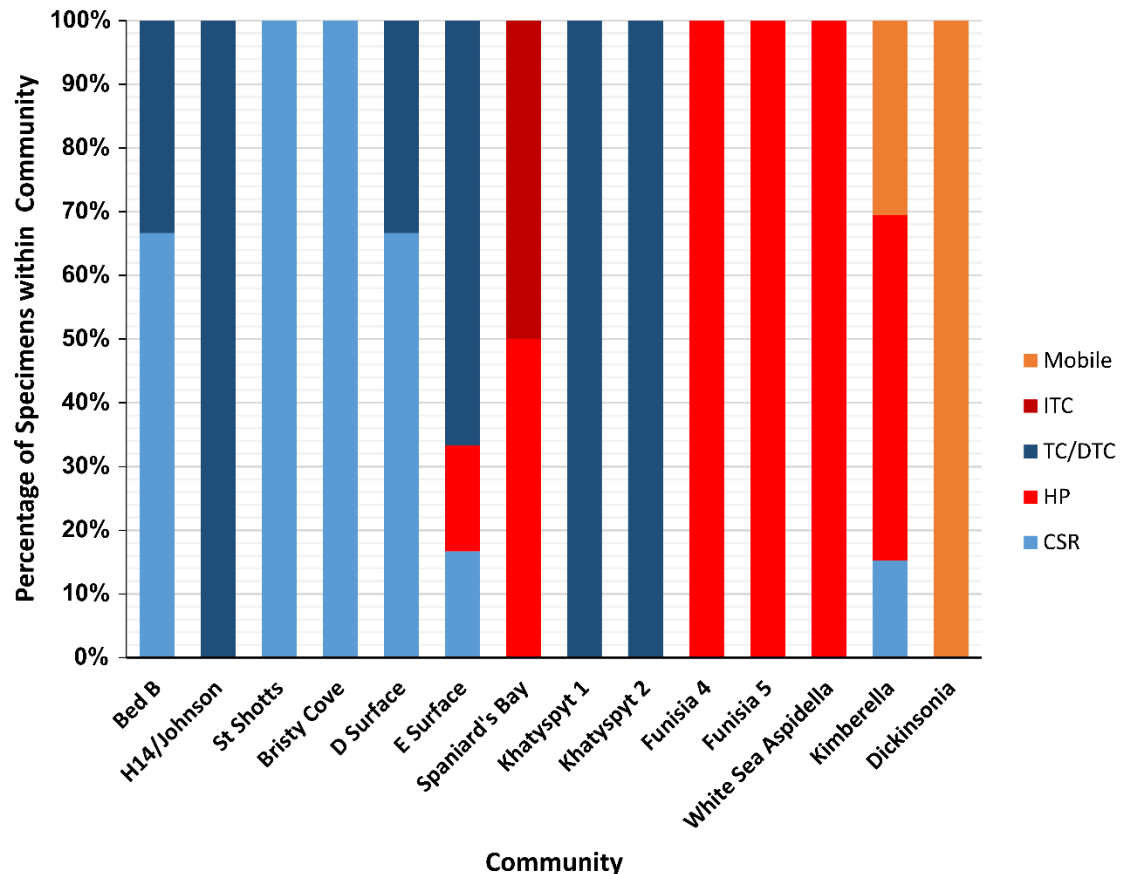


Fig. 4 Proportion of best-fit univariate models by surface, adapted from (23). The percentage of studied body-fossil specimens within the community with univariate spatial distributions that are best described by CSR, HP, TC (or DTC) and ITC models. CSR and TC are considered random or dispersal (neutral) models and are shown in blue. HP and ITC are local environmentally driven (niche) models, shown in red. Mobile taxa are shown in orange, and inferred to be environmentally-driven. Data and plot for surfaces Bed B to Spaniard's Bay from ref. .

Untangling environmental from evolutionary trends in the Ediacaran has been hampered by a limited overlap between temporal periods and environmental settings (1,17). The palaeocommunities in this study derive from successions containing a variety of lithologies (tuff, coarse sandstone, mixed siltstone, limestone) and reflecting different palaeogeographic

positions (17,57,64,65,76,106,107). We find no significant direct correlations between these factors and the relative importance of habitat heterogeneities on the studied surfaces ($p > 0.1$; Fig. 3, Table 2). The palaeocommunities that are not influenced by local habitat heterogeneities (KH1 and KH2) are hosted within carbonate successions (107), making them distinct from the siliciclastically-hosted palaeocommunities on the KS, WS-A, FUN4, FUN5 and DS surfaces, or in previous work (21–23). However, the Khatyspyt surfaces behave ecologically in the same way to Avalonian palaeocommunities derived from similar depths, but different lithological successions (21–23), suggesting that lithology alone may not be responsible for the KH1 and KH2 surfaces differing results. Therefore, two possible factors remain that may explain the differences in community dynamics found here. The differences could reflect evolutionary trends, and it is true that the oldest studied palaeocommunities show limited habitat influence (21–23) when compared to the younger palaeocommunities documented in this study (Fig. 4). Unfortunately, the lack of fine-scale dating across these communities and older Avalonian ones precludes detailed fine-scale regression to assess whether the Khatyspyt palaeocommunities are an outlier to this apparent trend, or this trend merely reflects the biases of the available data. Alternatively, the differences could be due to the environmental setting. We have shown that Ediacaran environmental setting has a significant influence on community dynamics ($p = 0.021$), with shallow water palaeocommunities significantly influenced by habitat heterogeneities, in contrast to the deep water palaeocommunities (Fig. 3, Table 2; 21–23). This result does not preclude a simultaneous temporal influence, but given our dataset we are limited to only assessing the broad environmental factor. Our dataset also precludes a more detailed analysis of the relationship between community ecologies with different environmental facies, hence the focus on broad signals (shallow/deep). The surfaces included in this study consist of a small proportion of the known Ediacaran palaeocommunities, and were frequently depauperate in taxonomic diversity,

so in order to fully corroborate the proposed hypotheses, analysis of more surfaces from a wider range of global localities is required.

While SPPA have only been applied to a small proportion of the known *in situ* Ediacaran palaeocommunities (17 studied surfaces to date), there is a notable correspondence between the importance of habitat heterogeneities to community ecology and assemblage diversity. In this study, the palaeocommunities exhibiting significant influence from local habitat heterogeneities belong to the diverse White Sea assemblage, in contrast to the previous work on Avalonian palaeocommunities (21–23), which are not significantly influenced by such heterogeneities. The relationship between environmental spatial heterogeneities and species richness is well established, with habitat variations enabling species to co-exist through the creation of different niches (108). This relationship extends to modern deep-sea benthic communities, where these heterogeneities have been shown to provide a mechanism for diversification on large scales, such as between canyons, trenches and seamounts (109,110), on the centimetre to metre scale (111), and through microhabitats (42).

In the modern oceans there are three relevant mechanisms that increase habitat heterogeneity, and they do so through increasing both substrate heterogeneity and variation in differentiated particulate organic carbon (POC) and matter (POM) within the water column. First, metazoan mat grazing creates substrate heterogeneity in microbial substrates through the formation of depleted and non-depleted patches (112). Secondly, this grazing induces creation of different-sized detrital particles in the form of differential-sized fecal pellets and fragments of non-consumed food within the water-column, leading to water-column heterogeneity (113). Differential POC and POM create new food sources and therefore new niches, and have been shown to increase the biodiversity of sessile benthic communities (114–116). Thirdly, the main

597 source of deep-sea habitat heterogeneity is small-scale variation due to differentiated particle
598 influx, with diurnal vertical migration of mesozooplankton and macrofauna contributing up to
599 ~50% of POC to the modern deep-sea via fecal pellets (117–119), with the remaining ~50%
600 transported from shallow to deep-water via oceanic currents (120). Taken together, these
601 modern processes describe how grazing in the shallow waters contributes to habitat
602 heterogeneity in deep-water communities in the form of differentiated POC/POM.

603

604 Tentatively, we propose that the ecological differentiation observed between Ediacaran
605 shallow and deep-water communities may evidence the late Ediacaran development of a similar
606 chain of evolutionary diversification. This chain started in shallow water communities, with
607 the creation of habitat patchiness by mobile Ediacaran organisms, which then led to a feedback
608 of increasing diversification that ultimately expanded into the deep-sea. This hypothesized
609 feedback could have promoted diversification through the latest Ediacaran by increasing
610 heterogeneity. First, our data suggest that once grazing had occurred, organisms such as
611 *Kimberella* may have avoided pre-grazed patches (Fig. 3C, Table 2), with this selective grazing
612 accelerating further creation of mat heterogeneity and water-column heterogeneity via
613 differentiated POC (cf. 111,112). Secondly, this differentiated POC and POM would have
614 been transported to the deep-sea via oceanic currents (120). Diurnal vertical migration was
615 likely absent (6) in the Ediacaran, because while a planktonic/larval stage for Ediacaran
616 organisms has been predicted on the basis of their likely waterborne dispersal strategies
617 (25,105), there is presently no direct evidence of non-larval, planktotrophic zooplankton until
618 the onset of the Cambrian . In the absence of planktotrophic zooplankton and widespread
619 pelagic macrofauna, the Ediacaran POC flux may have been either larger, due to lack of
620 consumption of phytoplankton in the shallow water, or smaller, due to a lack of mixing by
621 diurnal vertical migration of the plankton (6), and this cannot yet be determined. Prior to

622 grazers and detritivores, the POC/POM flux would have been dominated by relatively
623 homogenous phytodetritus. The evolution of grazers would have led to a shift towards size-
624 differentiated POC/POM, potentially increasing the heterogeneity of the deep-sea habitat via
625 water-column heterogenization (114–116), and so providing a mechanism for deep-marine
626 diversification.

627
628 Budd and Jensen (12) introduced the Savannah hypothesis to explain early animal
629 diversification, whereby early bilaterian diversification was driven by small-scale variations in
630 local habitat, primarily caused by the spatial distributions of sessile organisms. They argue that
631 it was the drive to find these heterogeneous distributed resources that led to novel evolutionary
632 innovations such as mobility. Our results demonstrate that at least some of the early animal
633 communities that contain mobile organisms were influenced by habitat variations, but the
634 limited number of studied surfaces means that we cannot test whether the observed patchiness
635 results from the spatial distributions of the sessile organisms, or another source. However, we
636 do describe a mechanism that links early animal diversification and benthic habitat patchiness
637 prior to the evolution of predators and wide-spread pelagic organisms. We show that taxa
638 such as *Kimberella* had a segregated distribution with trace fossils considered to be their
639 grazing traces (99), suggesting that they may have been capable of avoiding non-preferred
640 areas (possibly already consumed patches), revealing adaptation of behavior when interacting
641 with these patches. This adaptation theoretically has the capacity to drive further
642 diversification, initially dependent on the environmental-setting, starting in shallow water
643 depositional environments, and then, over time, moving into deeper water, but currently
644 available global fossil assemblages limit the testing of this prediction. If this hypothesis is
645 correct, we would expect deep-water assemblages to diversify during the terminal Ediacaran
646 and into the Cambrian. Our results therefore provide tentative support for the Savannah

hypothesis, suggesting that this late Ediacaran taxonomic diversification was a benthic event, which facilitated a chain of diversification by promoting marine habitat heterogeneities.

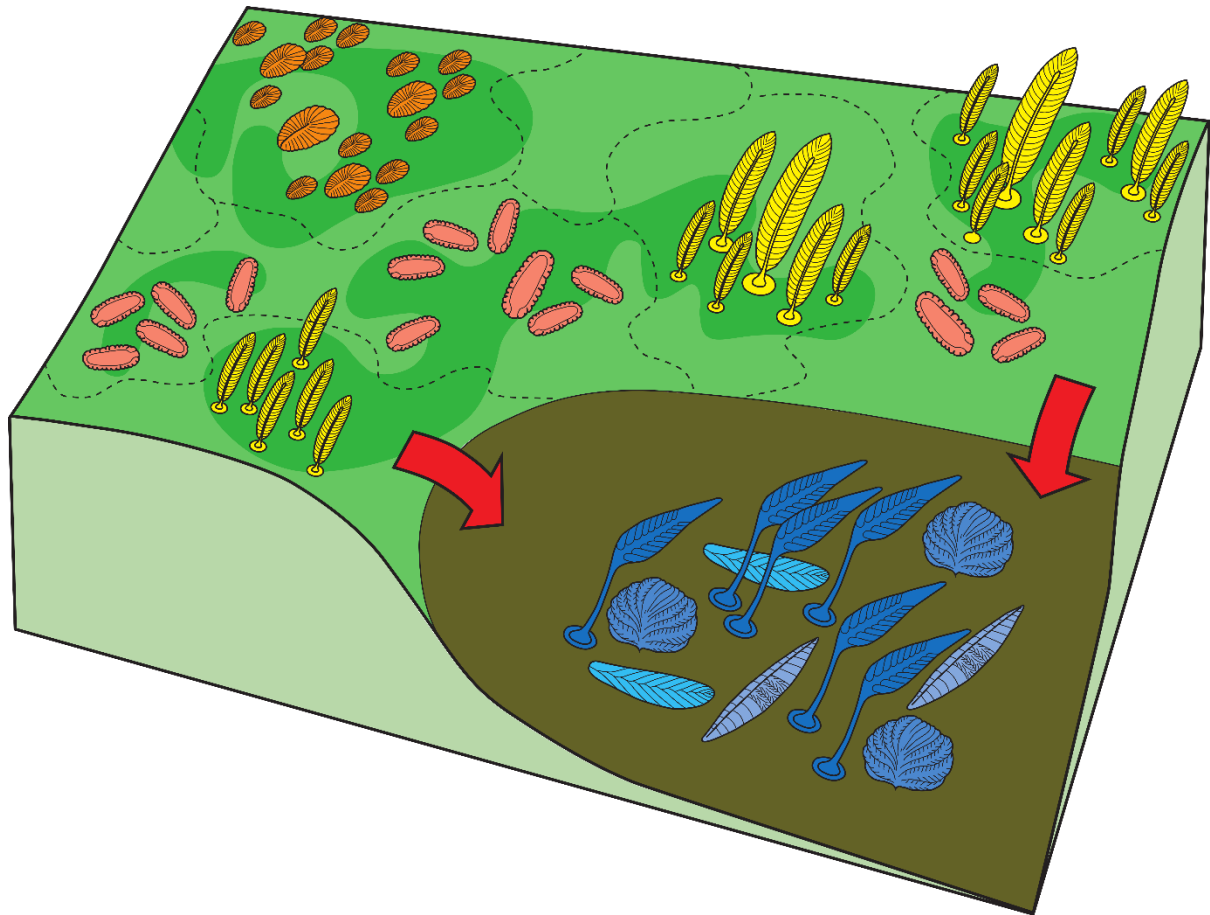


Fig. 5. Schematic diagram showing variation of heterogeneities within different environmental settings. Shallow water communities are significantly influenced by habitat heterogeneities. Grazing within these shallow waters further increases substrate heterogeneity, potentially increasing diversification. Furthermore, this grazing increases deep-water heterogeneity through the creation of different sized particulate organic matter due to the influx of particulate matter from the shallows, offering a potential mechanism via which to drive deep-marine diversification.

Conclusions

We present evidence to suggest that the influence of local habitat on Ediacaran organisms is significantly correlated with broad-scale environmental setting. The relationship of Ediacaran communities to habitat-dependent interactions is correlated with Ediacaran assemblage diversity, with communities from the more diverse White Sea assemblage showing significant habitat associations and interactions in contrast to relatively habitat insensitive deep-sea Avalonian assemblages. We suggest that the presence of shallow-water grazers could have created further habitat heterogeneity in shallow-water and ultimately deep-water, via the heterogenization of the shallow-water substrate and via the introduction of variable size particulate matter to the deep-sea. These results demonstrate the utility of these approaches for investigating the early diversification of metazoans. We have shown the importance of local environmental patchiness to the diversification of early animals, and our results are consistent with the hypothesis that the early diversification of metazoans was a benthic event, driven by responses to habitat patchiness.

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